# Mental Health & Incarceration

MCRT3 December 8, 2022

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### Agenda

01	Problem Statement
02	Data Exploration & Machine Learning
03	Limitations
04	Policy Recommendations & Future Work

### Problem Statement

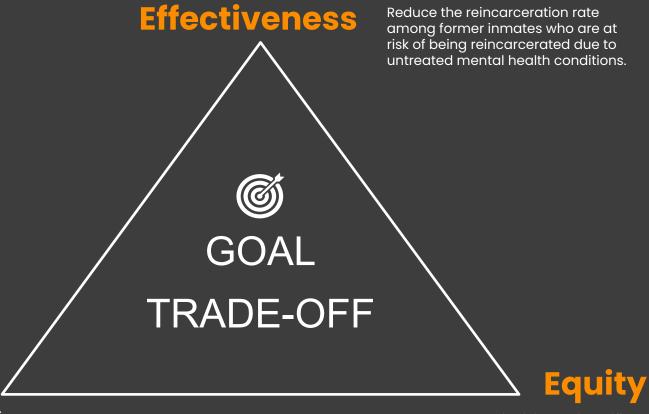
# More than 60% of inmates suffer from mental health conditions

### Goal

- Improve quality of life
- Positively impact community health
- Reduce rate of re-incarceration

### **Policy Impact**

- Identify 100 people who are most "at-risk" for re-incarceration
- Proactively provide mental health treatment



Due to constrained resources, we want to ensure that the 100 individuals are actually highest-risk of reincarcerated.

**Efficiency** 

Those 100 individuals identified monthly for proactive mental health outreach belong to communities most negatively affected by reincarceration.

# Data Exploration & Machine Learning

### **Data Sources**

01	Justice Information Management System Department (JIMS)	<b>)</b> :	Inmate Demographic and Background Information Case Information Pre-trial Risk Assessment Results
02	Johnson County Mental Health Center (JCMHC)	):	Financial Status Information  Mental Health Assessment Information  Crisis and Emergency Call Information
03	Johnson County Police Department	<b>)</b> :	Arrest Information Charge Information
04	Medical Emergencies Department (MEDACT)	•	Patient Encounter Information

On the first of every month, of all the individuals who were released from Johnson County Jail in the last 2 years and are not currently in jail, we aim to identify 100 individuals who are most likely to return to jail in the next year to prioritize proactive mental health interventions to prevent re-incarceration

### **Modelling Plan**

#### Rows

A person on a certain date

#### Labels

Whether or not this person will be re-incarcerated within the next year of this date.

The value is 1 if True, and 0 if False.

#### **Predictors**

Demographics | Criminal Justice History | Mental Health History | Risk Evaluation

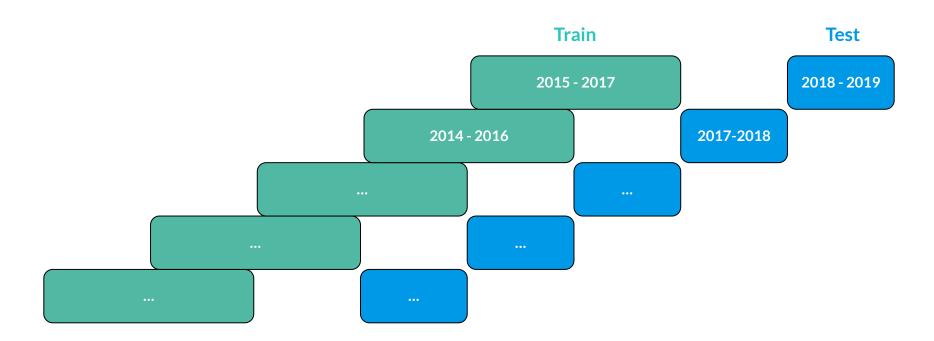
#### **Model Types**

Decision Tree | Random Forest | Scaled Logistic Regression | Gradient Boosting | Neural Network

#### **Model Selection Metrics**

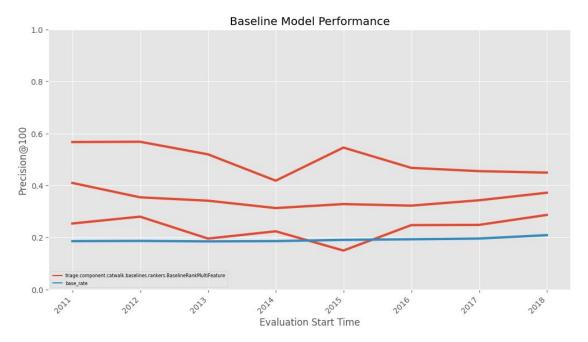
Out of the 100 people that the model assigned top risk scores, how many of them actually end up being reincarcerated.

### **Temporal Validation**

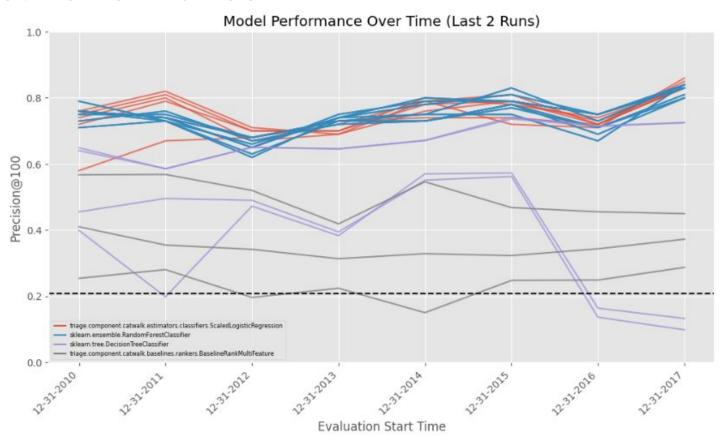


### **Baselines**

- 1. Number of bookings since the feature start time
- 2. People who have visited JCMHC the most since the feature start time
- 3. Average time between current release and next booking



### **Model Performance**

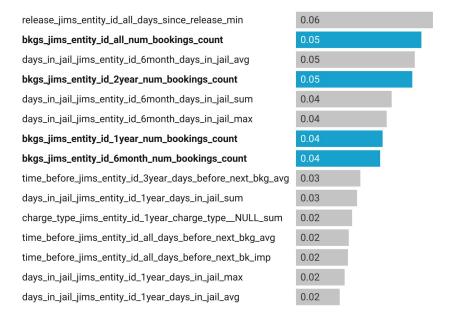


### Feature Importances

#### **Scaled Logistic Regression**

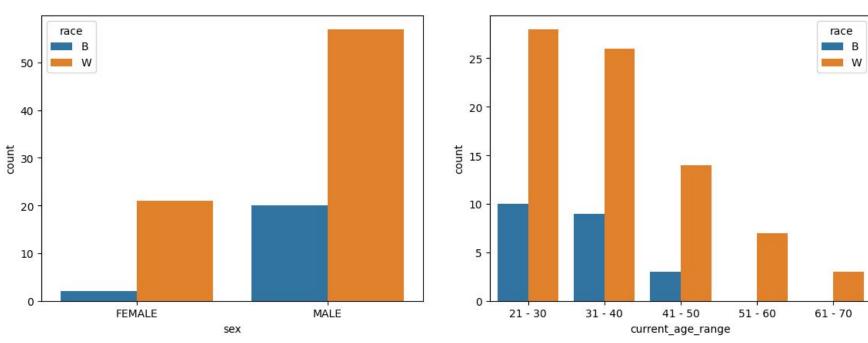
num_charges_pd_entity_id_2year_num_charges_count	2.7	
arrest_type_pd_entity_id_all_arr_type_S_sum	2.18	
charge_type_jims_entity_id_1year_charge_typeNULL_sum	1.93	
bkgs_jims_entity_id_all_num_bookings_count	1.88	
bkgs_jims_entity_id_6month_num_bookings_count	1.8	
bkgs_jims_entity_id_2year_num_bookings_count	1.56	
bkgs_jims_entity_id_1year_num_bookings_count	1.48	
pre_trial_ass_jims_entity_id_1year_felony_flag_max	1.29	
pre_trial_ass_jims_entity_id_2year_felony_flag_max	1.27	
demographics_jims_entity_id_all_marital_status_S_max	1.23	
charge_type_jims_entity_id_all_charge_typeNULL_sum	1.23	
demographics_jims_entity_id_all_sex_MALE_max	1.2	
pre_trial_ass_jims_entity_id_all_substance_flag_max	1.16	
$charge\_type\_jims\_entity\_id\_6month\_charge\_type\_\_NULL\_sum$	1.13	
bkgs_jims_entity_id_1month_num_bookings_count	1.12	

#### Random Forest



### Characteristics of Those Classified as At Risk

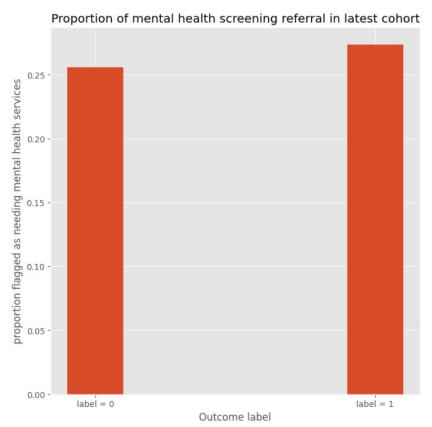




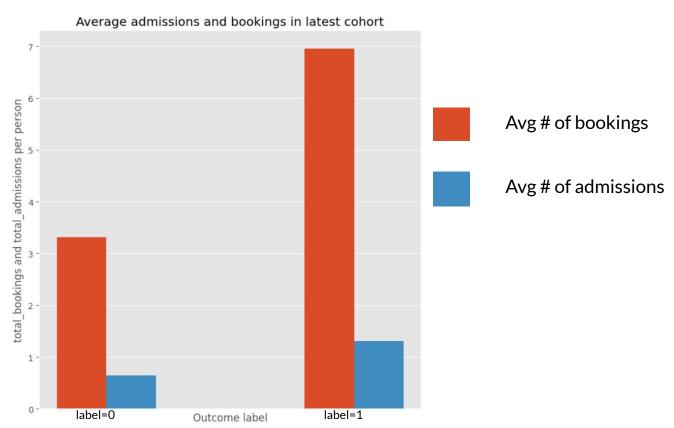
### Limitations

# Our analytical formulation does not explicitly incorporate mental health into the cohort or label structure

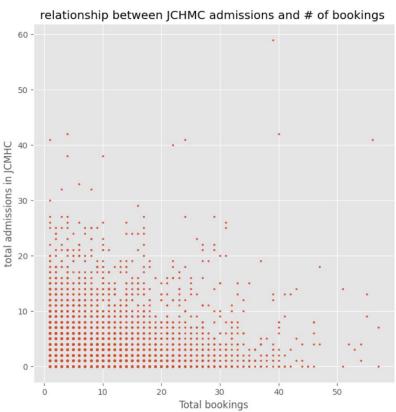
### There is a small difference between mental health outcomes in the latest cohort



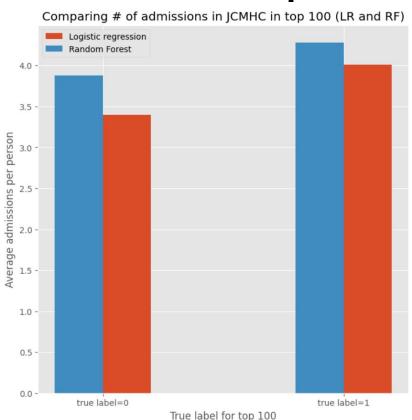
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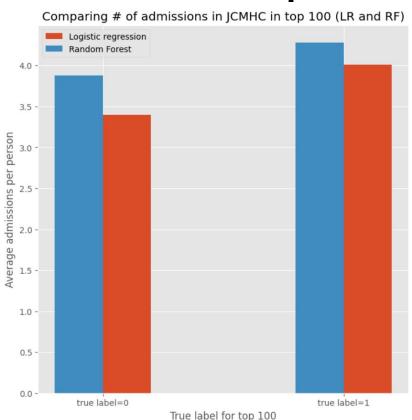
## We don't see a strong correlation between bookings and JCMHC admissions



# There is a small difference in MH outcomes between correct and incorrect predictions



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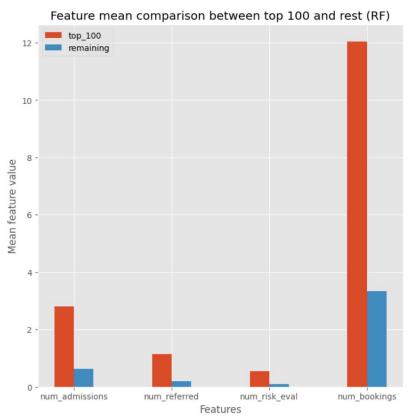
### How important are the mental health features?



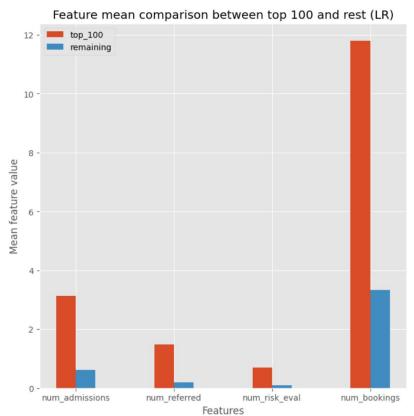
Best rank any feature gets in any mode  $\rightarrow$  8

Best ranked MH feature  $\rightarrow$  # of admissions in JCMHC

## Criminal justice features have a higher difference between top 100 and the rest



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### Policy Recommendations & Future Work

# Targeting interventions

- GAO: Importance of early screening
  - Expansion of BJMHS
  - Follow-up surveying
  - Assess survey for potential bias
- Proactivity is key
- Personalized care plans, during and after incarceration

"Individuals should be screened as early in the booking process as possible and throughout their involvement in the criminal justice system to detect substance abuse and mental health disorders."

-GAO

### **Future Work**

### Incorporating mental health conditions in cohort definition

- Not currently accounted for in our definition
- May impact usefulness of treatment
- Expansion of BJMHS helpful in this

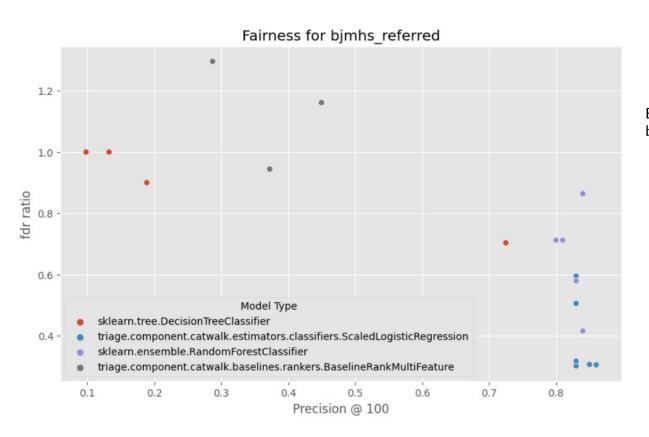
#### Field trial deployment

- Measuring efficacy of outreach
- Measuring potentially deleterious effects of outreach
- Regression discontinuity at 100 people served
- Iteration

### **Questions?**

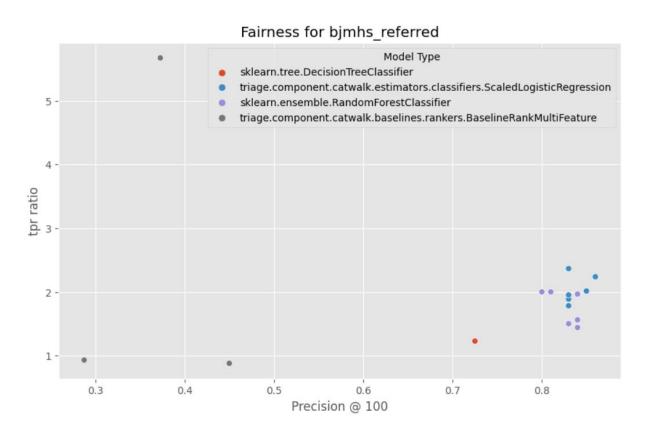
### Appendix

### Quick note on fairness performance on MH feature



Base group: bjmhs\_referred = 0

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